

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
from google.colab import files
uploaded = files.upload()
```

Choose Files

House Price India.csv

- House Price India.csv(text/csv) - 1524561 bytes, last modified: 9/26/2023 - 100% done

Saving House Price India.csv to House Price India.csv

```
import io
df = pd.read_csv(io.BytesIO(uploaded['House Price India.csv']))
```

```
df.head()
```

	id	Date	number of bedrooms	number of bathrooms	living area	lot area	number of floors	waterfront present	number of views	condition of the house	...	Buil Yea
0	6762810145	42491	5	2.50	3650	9050	2.0	0	4	5	...	192
1	6762810635	42491	4	2.50	2920	4000	1.5	0	0	5	...	190
2	6762810998	42491	5	2.75	2910	9480	1.5	0	0	3	...	193
3	6762812605	42491	4	2.50	3310	42998	2.0	0	0	3	...	200
4	6762812919	42491	3	2.00	2710	4500	1.5	0	0	4	...	192

5 rows × 23 columns

```
df.tail()
```

	id	Date	number of bedrooms	number of bathrooms	living area	lot area	number of floors	waterfront present	number of views	condition of the house	...
14615	6762830250	42734	2	1.5	1556	20000	1.0	0	0	4	...
14616	6762830339	42734	3	2.0	1680	7000	1.5	0	0	4	...
14617	6762830618	42734	2	1.0	1070	6120	1.0	0	0	3	...
14618	6762830709	42734	4	1.0	1030	6621	1.0	0	0	4	...
14619	6762831463	42734	3	1.0	900	4770	1.0	0	0	3	...

5 rows × 23 columns

```
df
```

```

      id  Date      number of bedrooms  number of bathrooms  living area  lot area  number of floors  waterfront present  number of views  condition of the house  ...
df.columns

Index(['id', 'Date', 'number of bedrooms', 'number of bathrooms',
      'living area', 'lot area', 'number of floors', 'waterfront present',
      'number of views', 'condition of the house', 'grade of the house',
      'Area of the house(excluding basement)', 'Area of the basement',
      'Built Year', 'Renovation Year', 'Postal Code', 'Latitude',
      'Longitude', 'living_area_renov', 'lot_area_renov',
      'Number of schools nearby', 'Distance from the airport', 'Price'],
      dtype='object')

```

```
df.dtypes
```

```

id                int64
Date              int64
number of bedrooms    int64
number of bathrooms  float64
living area          int64
lot area            int64
number of floors      float64
waterfront present    int64
number of views       int64
condition of the house int64
grade of the house    int64
Area of the house(excluding basement) int64
Area of the basement  int64
Built Year           int64
Renovation Year       int64
Postal Code          int64
Latitude             float64
Longitude            float64
living_area_renov     int64
lot_area_renov        int64
Number of schools nearby int64
Distance from the airport int64
Price               int64
dtype: object

```

```
df.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14620 entries, 0 to 14619
Data columns (total 23 columns):
 #   Column                                Non-Null Count  Dtype
---  -
 0   id                                    14620 non-null  int64
 1   Date                                14620 non-null  int64
 2   number of bedrooms                  14620 non-null  int64
 3   number of bathrooms                 14620 non-null  float64
 4   living area                         14620 non-null  int64
 5   lot area                           14620 non-null  int64
 6   number of floors                    14620 non-null  float64
 7   waterfront present                  14620 non-null  int64
 8   number of views                     14620 non-null  int64
 9   condition of the house              14620 non-null  int64
10   grade of the house                  14620 non-null  int64
11   Area of the house(excluding basement) 14620 non-null  int64
12   Area of the basement                14620 non-null  int64
13   Built Year                          14620 non-null  int64
14   Renovation Year                     14620 non-null  int64
15   Postal Code                         14620 non-null  int64
16   Latitude                           14620 non-null  float64
17   Longitude                           14620 non-null  float64
18   living_area_renov                   14620 non-null  int64
19   lot_area_renov                     14620 non-null  int64
20   Number of schools nearby             14620 non-null  int64
21   Distance from the airport           14620 non-null  int64
22   Price                              14620 non-null  int64
dtypes: float64(4), int64(19)
memory usage: 2.6 MB

```

```
df.shape
```

```
(14620, 23)
```

```
print(df.describe())
```

```

std      6.237575e+03      67.347991      0.938719      0.769934

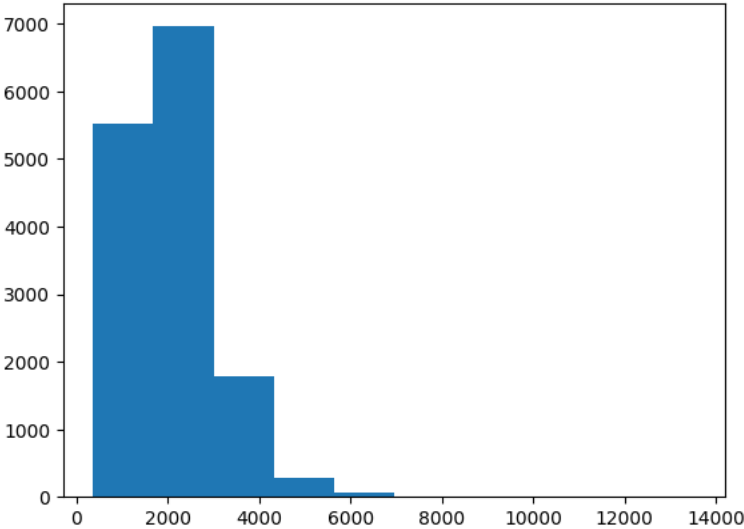
```

/3%	0.102620e+03	4.2052.000000	4.000000	2.500000
max	6.762832e+09	42734.000000	33.000000	8.000000
	living area	lot area	number of floors	waterfront present \
count	14620.000000	1.462000e+04	14620.000000	14620.000000
mean	2098.262996	1.509328e+04	1.502360	0.007661
std	928.275721	3.791962e+04	0.540239	0.087193
min	370.000000	5.200000e+02	1.000000	0.000000
25%	1440.000000	5.010750e+03	1.000000	0.000000
50%	1930.000000	7.620000e+03	1.500000	0.000000
75%	2570.000000	1.080000e+04	2.000000	0.000000
max	13540.000000	1.074218e+06	3.500000	1.000000
	number of views	condition of the house ...	Built Year \	
count	14620.000000	14620.000000 ...	14620.000000	
mean	0.233105	3.430506 ...	1970.926402	
std	0.766259	0.664151 ...	29.493625	
min	0.000000	1.000000 ...	1900.000000	
25%	0.000000	3.000000 ...	1951.000000	
50%	0.000000	3.000000 ...	1975.000000	
75%	0.000000	4.000000 ...	1997.000000	
max	4.000000	5.000000 ...	2015.000000	
	Renovation Year	Postal Code	Lattitude	Longitude \
count	14620.000000	14620.000000	14620.000000	14620.000000
mean	90.924008	122033.062244	52.792848	-114.404007
std	416.216661	19.082418	0.137522	0.141326
min	0.000000	122003.000000	52.385900	-114.709000
25%	0.000000	122017.000000	52.707600	-114.519000
50%	0.000000	122032.000000	52.806400	-114.421000
75%	0.000000	122048.000000	52.908900	-114.315000
max	2015.000000	122072.000000	53.007600	-113.505000
	living_area_renov	lot_area_renov	Number of schools nearby \	
count	14620.000000	14620.000000	14620.000000	
mean	1996.702257	12753.500068	2.012244	
std	691.093366	26058.414467	0.817284	
min	460.000000	651.000000	1.000000	
25%	1490.000000	5097.750000	1.000000	
50%	1850.000000	7620.000000	2.000000	
75%	2380.000000	10125.000000	3.000000	
max	6110.000000	560617.000000	3.000000	
	Distance from the airport	Price		
count	14620.000000	1.462000e+04		
mean	64.950958	5.389322e+05		
std	8.936008	3.675324e+05		
min	50.000000	7.800000e+04		
25%	57.000000	3.200000e+05		
50%	65.000000	4.500000e+05		
75%	73.000000	6.450000e+05		
max	80.000000	7.700000e+06		

[8 rows x 23 columns]

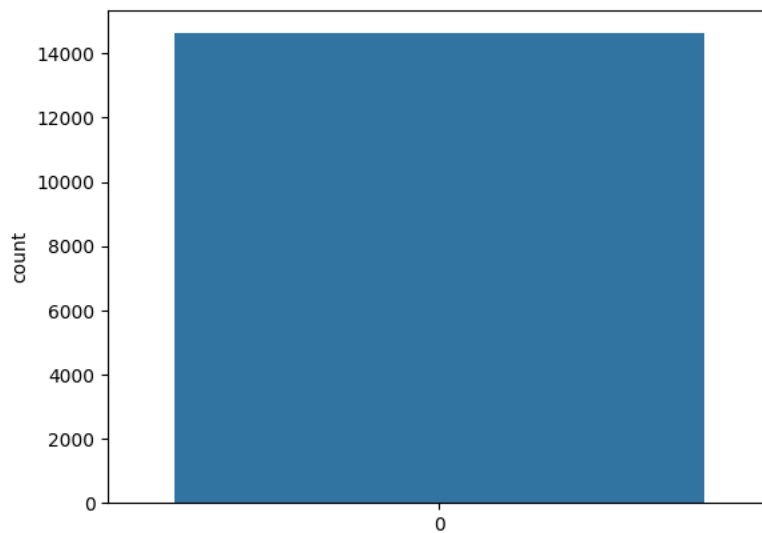
```
plt.hist(df['living area'])
```

```
(array([5.530e+03, 6.958e+03, 1.779e+03, 2.770e+02, 5.500e+01, 1.400e+01,
        2.000e+00, 3.000e+00, 1.000e+00, 1.000e+00]),
array([ 370., 1687., 3004., 4321., 5638., 6955., 8272., 9589.,
        10906., 12223., 13540.]),
<BarContainer object of 10 artists>)
```



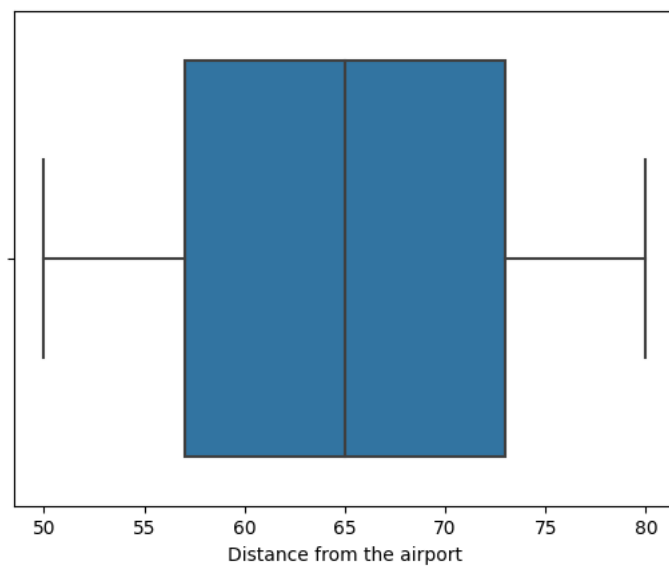
```
sns.countplot(df['condition of the house'])
```

<Axes: ylabel='count'>



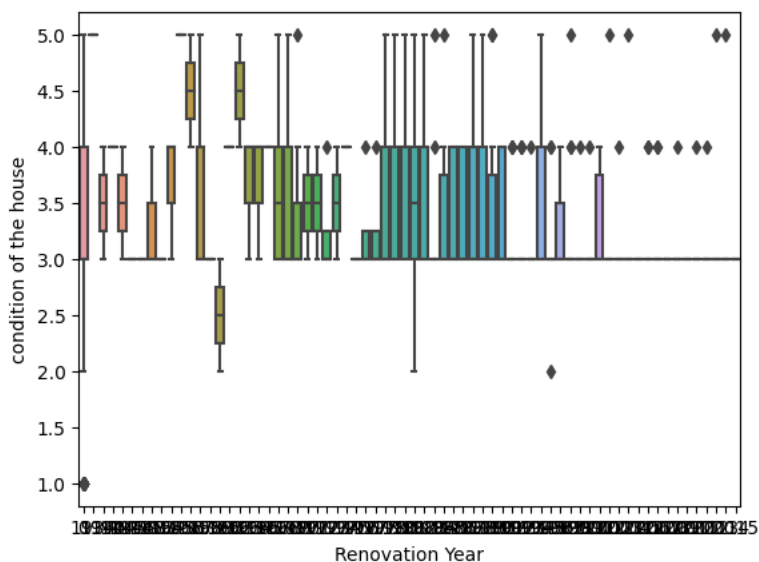
```
sns.boxplot(x=df['Distance from the airport'])
```

<Axes: xlabel='Distance from the airport'>



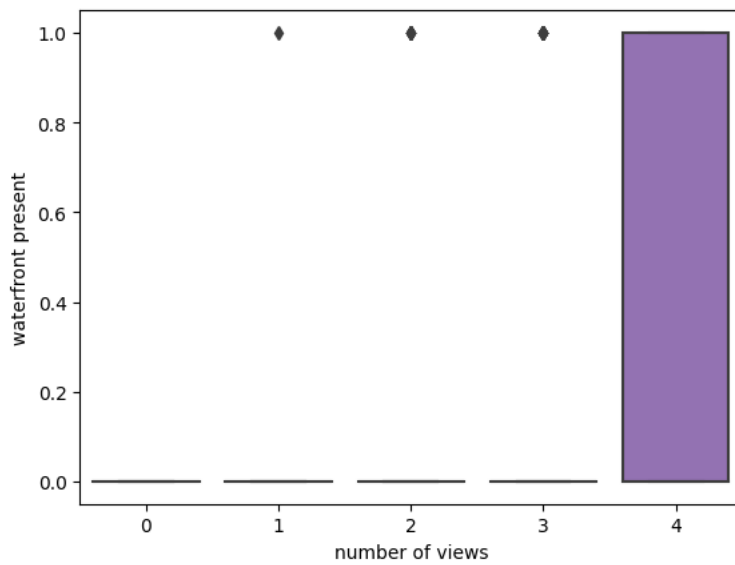
```
sns.boxplot(x=df['Renovation Year'], y=df['condition of the house'])
```

<Axes: xlabel='Renovation Year', ylabel='condition of the house'>



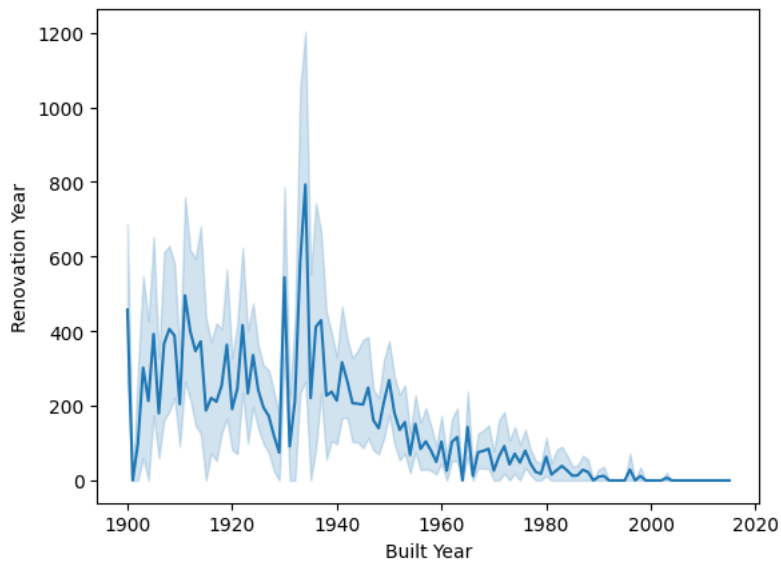
```
sns.boxplot(x=df['number of views'], y=df['waterfront present'])
```

<Axes: xlabel='number of views', ylabel='waterfront present'>



```
sns.lineplot(x=df['Built Year'],y=df['Renovation Year'])
```

<Axes: xlabel='Built Year', ylabel='Renovation Year'>



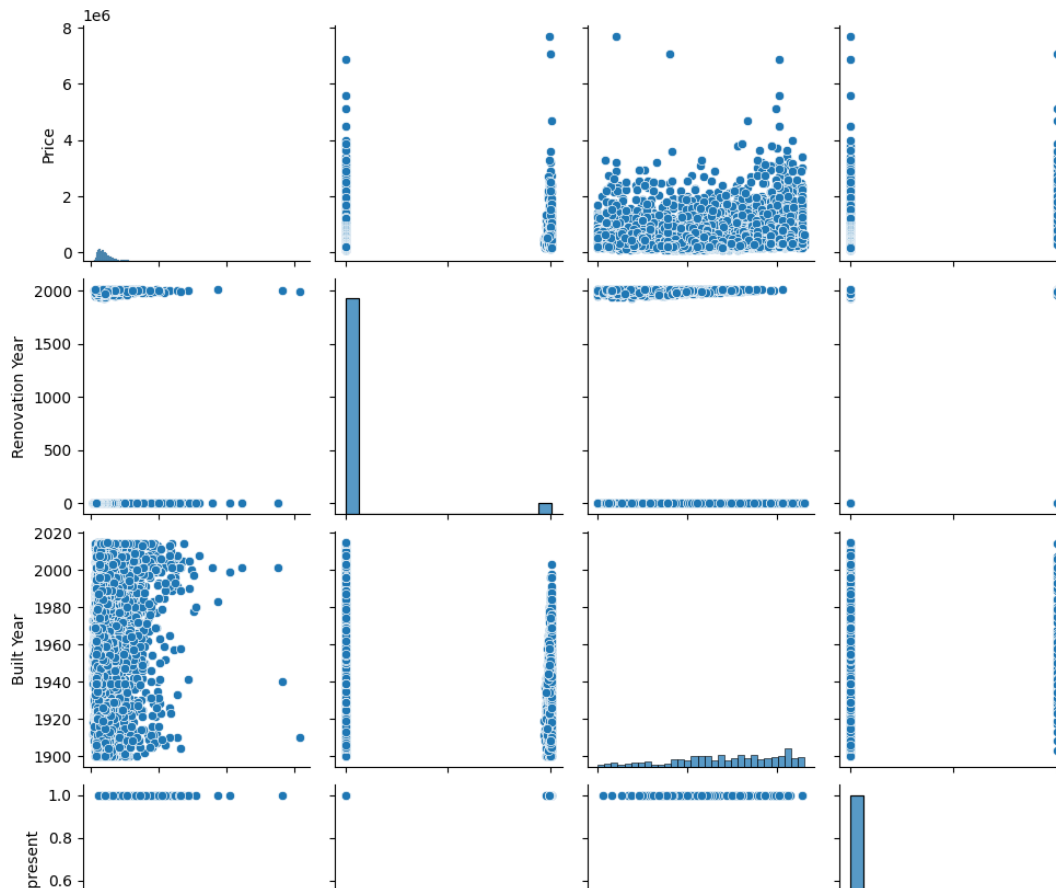
```
sns.heatmap(df[['living_area_renov','lot_area_renov','living area']].corr(),annot=T
```

<Axes: >

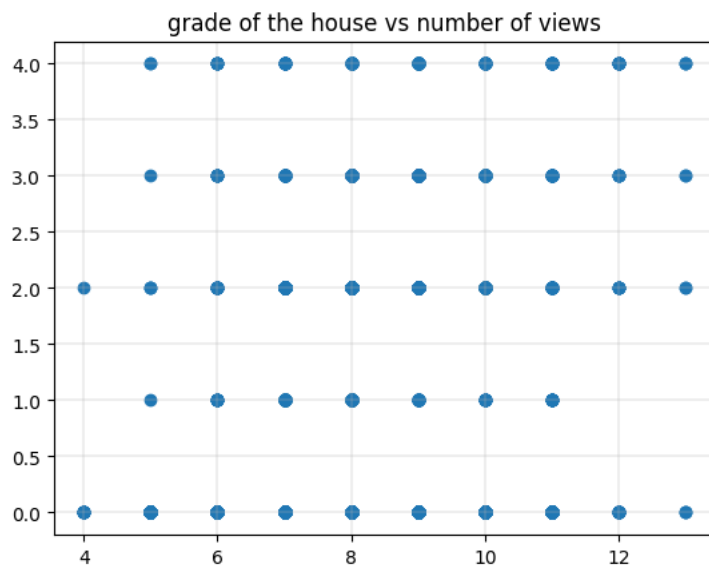


```
sns.pairplot(df[['Price', 'Renovation Year', 'Built Year', 'waterfront present']])
```

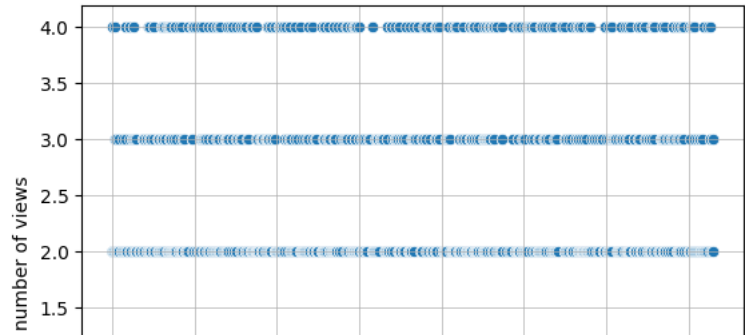
```
<seaborn.axisgrid.PairGrid at 0x7ddf24017fd0>
```



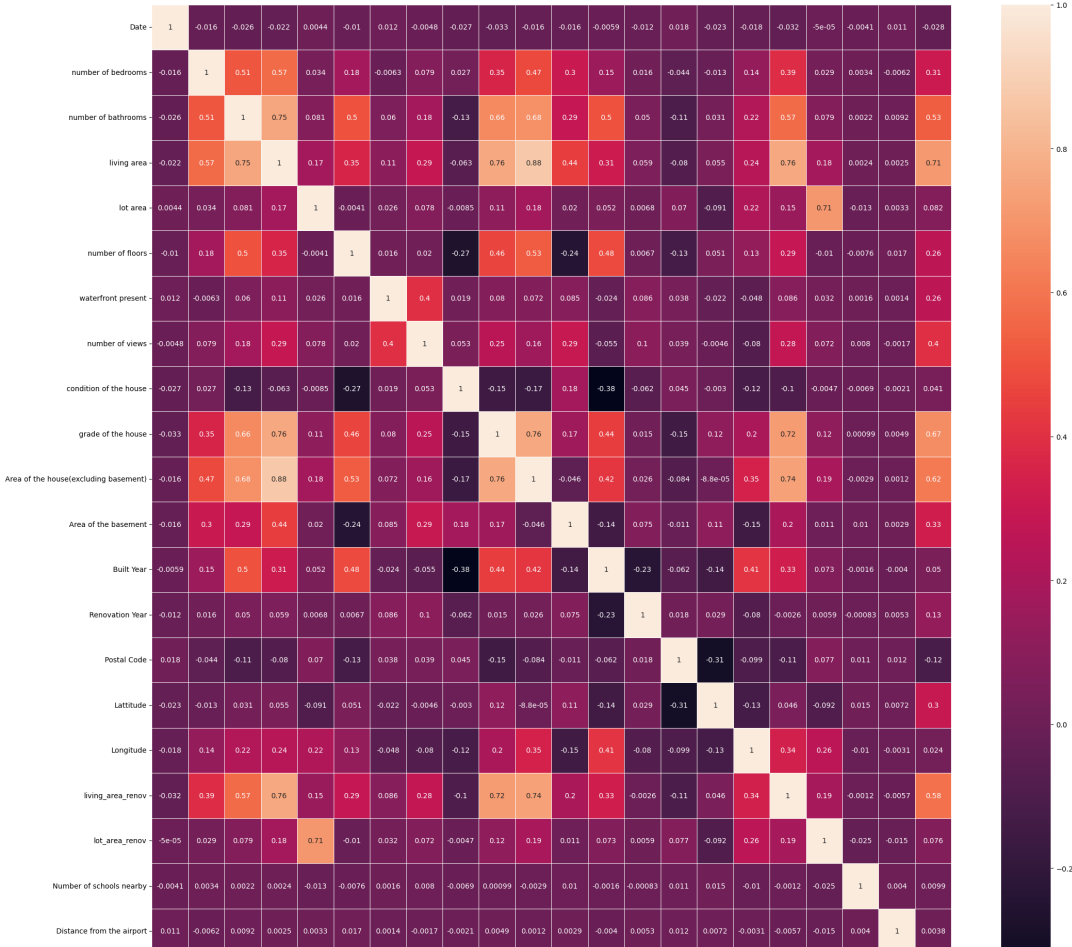
```
plt.scatter(df['grade of the house'],df['number of views'])
plt.title("grade of the house vs number of views")
plt.grid(linestyle='-', linewidth=0.3)
```



```
sns.scatterplot(df['number of views'])
plt.grid(linestyle='-', linewidth=0.5)
```



```
plt.subplots(figsize=(25,25))
sns.heatmap(df.drop(['id'],axis=1).corr(),linewidth=0.4,annot=True)
plt.show()
```



```
print(df.describe())
```

std	6.237575e+03	67.347991	0.938719	0.769934
min	6.762810e+09	42491.000000	1.000000	0.500000
25%	6.762815e+09	42546.000000	3.000000	1.750000
50%	6.762821e+09	42600.000000	3.000000	2.250000
75%	6.762826e+09	42662.000000	4.000000	2.500000
max	6.762832e+09	42734.000000	33.000000	8.000000

	living area	lot area	number of floors	waterfront present \
count	14620.000000	1.462000e+04	14620.000000	14620.000000
mean	2098.262996	1.509328e+04	1.502360	0.007661
std	928.275721	3.791962e+04	0.540239	0.087193
min	370.000000	5.200000e+02	1.000000	0.000000
25%	1440.000000	5.010750e+03	1.000000	0.000000
50%	1930.000000	7.620000e+03	1.500000	0.000000
75%	2570.000000	1.080000e+04	2.000000	0.000000
max	13540.000000	1.074218e+06	3.500000	1.000000

	number of views	condition of the house	...	Built Year \
count	14620.000000	14620.000000	...	14620.000000
mean	0.233105	3.430506	...	1970.926402
std	0.766259	0.664151	...	29.493625
min	0.000000	1.000000	...	1900.000000

```

max      4.000000      5.000000 ...      2015.000000

Renovation Year      Postal Code      Latitude      Longitude \
count      14620.000000      14620.000000      14620.000000      14620.000000
mean       90.924008      122033.062244      52.792848      -114.404007
std        416.216661      19.082418      0.137522      0.141326
min         0.000000      122003.000000      52.385900      -114.709000
25%         0.000000      122017.000000      52.707600      -114.519000
50%         0.000000      122032.000000      52.806400      -114.421000
75%         0.000000      122048.000000      52.908900      -114.315000
max        2015.000000      122072.000000      53.007600      -113.505000

living_area_renov      lot_area_renov      Number of schools nearby \
count      14620.000000      14620.000000      14620.000000
mean       1996.702257      12753.500068      2.012244
std        691.093366      26058.414467      0.817284
min         460.000000      651.000000      1.000000
25%         1490.000000      5097.750000      1.000000
50%         1850.000000      7620.000000      2.000000
75%         2380.000000      10125.000000      3.000000
max         6110.000000      560617.000000      3.000000

Distance from the airport      Price
count      14620.000000      1.462000e+04
mean       64.950958      5.389322e+05
std         8.936008      3.675324e+05
min         50.000000      7.800000e+04
25%         57.000000      3.200000e+05
50%         65.000000      4.500000e+05
75%         73.000000      6.450000e+05
max         80.000000      7.700000e+06

```

[8 rows x 23 columns]

```
print(df.count())
```

```

id      14620
Date    14620
number of bedrooms    14620
number of bathrooms   14620
living area           14620
lot area              14620
number of floors      14620
waterfront present    14620
number of views       14620
condition of the house 14620
grade of the house     14620
Area of the house(excluding basement) 14620
Area of the basement   14620
Built Year            14620
Renovation Year        14620
Postal Code            14620
Latitude              14620
Longitude             14620
living_area_renov      14620
lot_area_renov         14620
Number of schools nearby 14620
Distance from the airport 14620
Price                 14620
dtype: int64

```

```
print(df.corr())
```



```

Latitude          0.014949
Longitude         -0.010163
living_area_renov -0.001203
lot_area_renov    -0.025014
Number of schools nearby 1.000000
Distance from the airport 0.004035
Price             0.009890

```

```

                Distance from the airport    Price
id          -0.004542 -0.773114
Date          0.011457 -0.027919
number of bedrooms -0.006157  0.308460
number of bathrooms  0.009206  0.531735
living area    0.002511  0.712169
lot area       0.003291  0.081992
number of floors  0.016567  0.262732
waterfront present  0.001448  0.263687
number of views  -0.001657  0.395973
condition of the house -0.002136  0.041376
grade of the house    0.004940  0.671814
Area of the house(excluding basement)  0.001222  0.615220
Area of the basement  0.002926  0.330202
Built Year          -0.003968  0.050307
Renovation Year      0.005342  0.133173
Postal Code          0.011528 -0.115908
Latitude             0.007193  0.297490
Longitude            -0.003100  0.024414
living_area_renov    -0.005673  0.584924
lot_area_renov       -0.014587  0.075535
Number of schools nearby  0.004035  0.009890
Distance from the airport 1.000000  0.003804
Price                0.003804  1.000000

```

```
[23 rows x 23 columns]
```

```
print(df['number of views'].value_counts())
```

```

0      13198
2         636
3         351
1         219
4         216
Name: number of views, dtype: int64

```

```

print('Mean:',df['Number of schools nearby'].mean())
print('Median:',df['Distance from the airport'].median())
print('Mode:',df['Postal Code'].mode())

```

```

Mean: 2.0122435020519838
Median: 65.0
Mode: 0      122028
Name: Postal Code, dtype: int64

```

```
print(df.isnull().sum())
```

```

id          0
Date        0
number of bedrooms  0
number of bathrooms  0
living area    0
lot area       0
number of floors  0
waterfront present  0
number of views  0
condition of the house  0
grade of the house    0
Area of the house(excluding basement)  0
Area of the basement  0
Built Year        0
Renovation Year    0
Postal Code        0
Latitude          0
Longitude         0
living_area_renov  0
lot_area_renov     0
Number of schools nearby  0
Distance from the airport  0
Price            0
dtype: int64

```

```

df.dropna(inplace=True)
df.fillna(0,inplace=True)
df.interpolate(inplace=True)

```

```
from sklearn.preprocessing import StandardScaler
from sklearn.preprocessing import MinMaxScaler
x=df.drop(['Price','Date'],axis=1)
x.set_index(['id'],inplace=True)
y=df[['id','Price']]
```

```
x.head()
```

	number of bedrooms	number of bathrooms	living area	lot area	number of floors	waterfront present	number of views	condition of the house	grade of the house	Area of house(exclud baseme
id										
6762810145	5	2.50	3650	9050	2.0	0	4	5	10	3
6762810635	4	2.50	2920	4000	1.5	0	0	5	8	1
6762810998	5	2.75	2910	9480	1.5	0	0	3	8	2
6762812605	4	2.50	3310	42998	2.0	0	0	3	9	3
6762812919	3	2.00	2710	4500	1.5	0	0	4	8	1

```
y.head()
```

	id	Price
0	6762810145	2380000
1	6762810635	1400000
2	6762810998	1200000
3	6762812605	838000
4	6762812919	805000

```
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestRegressor
from sklearn.ensemble import GradientBoostingRegressor
from sklearn.metrics import r2_score
x_train,x_test,y_train,y_test = train_test_split(x,y['Price'],test_size =0.1,random_state=2)
model = GradientBoostingRegressor(n_estimators=400,max_depth=5,min_samples_split=2,learning_rate=0.1)
model.fit(x_train,y_train)
```

▼ GradientBoostingRegressor

GradientBoostingRegressor(max_depth=5, n_estimators=400)

```
y_pred = model.predict(x_test)
model.score(x_test,y_test)
```

0.9115605737002777

```
r2_score(y_pred,y_test)
```



0.9008207044846054

```
y_pred
```

array([497766.12740438, 244495.3776842 , 293819.40063242, ...,
698495.60350629, 297006.00386358, 245881.76921871])

```
y_pred_list = y['id'][-len(y_pred):].tolist()
y_pred_df=pd.DataFrame(y_pred_list,columns=['ID'])
y_pred_df['Predicted Price']= y_pred.round(2)
```

```
y_pred_df
```

	ID	Predicted Price	
0	6762811233	497766.13	
1	6762811403	244495.38	
2	6762811775	293819.40	
3	6762811861	397555.35	
4	6762812009	474843.29	
...	
1457	6762830250	1041014.57	
1458	6762830339	317512.59	
1459	6762830618	698495.60	
1460	6762830709	297006.00	
1461	6762831463	245881.77	

1462 rows × 2 columns